AMOC and water-mass transformation in high- and low-resolution models: Climatology and variability

Dylan Charles Shamban Oldenburg¹, Robert C. J. Wills¹, Kyle Armour¹, and LuAnne Thompson¹

¹University of Washington

November 30, 2022

Abstract

Water-mass transformation in the North Atlantic plays an important role in the Atlantic Meridional Overturning Circulation (AMOC) and its variability. Here we analyze subpolar North Atlantic water-mass transformation in high- and low-resolution versions of the Community Earth System Model (CESM1) and investigate whether differences in resolution and climatological water-mass transformation impact low-frequency AMOC variability. We find that high-resolution simulations reproduce the water-mass transformation found in a reanalysis-forced high-resolution ocean simulation more accurately than low-resolution simulations. We also find that the low-resolution CESM1 simulations, including one forced with the same atmospheric reanalysis data, have larger biases in surface heat fluxes, sea-surface temperatures and salinities compared to the high-resolution simulations. Despite these major climatological differences, the mechanisms of low-frequency AMOC variability are similar in the high- and low-resolution versions of CESM1. The Labrador Sea WMT plays a major role in driving AMOC variability, and a similar NAO-like sea-level pressure pattern leads AMOC changes. However, the high-resolution simulation shows a more pronounced atmospheric response to the AMOC variability. The consistent role of Labrador Sea WMT in low-frequency AMOC variability across high- and low-resolution coupled simulations, including a simulation which accurately reproduces the WMT found in an atmospheric reanalysis-forced high-resolution ocean simulation, suggests that the mechanisms are similar in the real world.

AMOC and water-mass transformation in high- and low-resolution models: Climatology and variability

Dylan Oldenburg¹, Robert C. J. Wills², Kyle C. Armour^{1,2}, LuAnne Thompson¹

¹School of Oceanography, University of Washington, Seattle, Washington ²Department of Atmospheric Sciences, University of Washington, Seattle, Washington

Key Points:

1

2

3

4

5

6

7	• A high-resolution coupled simulation reproduces subpolar North Atlantic water-
8	mass transformation from a reanalysis-forced ocean simulation
9	Low-resolution simulations have larger biases in sea-surface heat fluxes, tem-
10	perature and salinity than the high-resolution simulations
11	• Despite climatological differences between the low- and high-resolution mod-
12	els, mechanisms of low-frequency AMOC variability are similar

Corresponding author: Dylan Oldenburg, oldend@uw.edu

13 Abstract

Water-mass transformation in the North Atlantic plays an important role in the At-14 lantic Meridional Overturning Circulation (AMOC) and its variability. Here we an-15 alyze subpolar North Atlantic water-mass transformation in high- and low-resolution 16 versions of the Community Earth System Model (CESM1) and investigate whether dif-17 ferences in resolution and climatological water-mass transformation impact low-frequency 18 AMOC variability. We find that high-resolution simulations reproduce the water-mass 19 transformation found in a reanalysis-forced high-resolution ocean simulation more ac-20 curately than low-resolution simulations. We also find that the low-resolution CESM1 21 simulations, including one forced with the same atmospheric reanalysis data, have larger 22 biases in surface heat fluxes, sea-surface temperatures and salinities compared to the 23 high-resolution simulations. Despite these major climatological differences, the mech-24 anisms of low-frequency AMOC variability are similar in the high- and low-resolution 25 versions of CESM1. The Labrador Sea WMT plays a major role in driving AMOC vari-26 ability, and a similar NAO-like sea-level pressure pattern leads AMOC changes. How-27 ever, the high-resolution simulation shows a more pronounced atmospheric response 28 to the AMOC variability. The consistent role of Labrador Sea WMT in low-frequency 29 AMOC variability across high- and low-resolution coupled simulations, including a 30 simulation which accurately reproduces the WMT found in an atmospheric reanalysis-31 forced high-resolution ocean simulation, suggests that the mechanisms are similar in 32 the real world. 33

³⁴ Plain Language Summary

Water-mass transformation, which refers to the process of converting a water par-35 cel from one density to another, plays an important role in the Atlantic Meridional Over-36 turning Circulation (AMOC). Here we use high- and low-resolution climate models 37 to investigate whether differences in the model resolution and time-mean water-mass 38 transformation patterns impact AMOC fluctuations. We find that high-resolution cou-39 pled simulations reproduce the water-mass transformation found in a high-resolution 40 ocean simulation driven by atmospheric reanalysis data, which we take as our clos-41 est analogue to observations. We also find that the low-resolution simulations, includ-42 ing one forced with the same atmospheric reanalyses, have larger discrepancies in sur-43 face properties compared to the high-resolution coupled simulation. Despite these ma-44

jor differences, the mechanisms driving AMOC variations are similar in the high and 45 low-resolution coupled simulations. Changes in water-mass transformation in the Labrador 46 Sea play a major role in driving AMOC fluctuations, and a similar sea-level pressure 47 pattern leads AMOC changes. The consistent role of the Labrador Sea in AMOC vari-48 ations across high- and low-resolution coupled simulations, including a high-resolution 49 simulation which accurately reproduces the water-mass transformation patterns found 50 in our closest analogue to observations, suggests that these mechanisms are similar 51 in the real world. 52

53 1 Introduction

The Atlantic Meridional Overturning Circulation (AMOC) plays an important 54 role in global climate by transporting large amounts of heat northward into the high 55 latitudes. The North Atlantic Current, which forms the upper branch of AMOC, car-56 ries warm, salty subtropical water northwards into the subpolar regions, releasing large 57 amounts of heat to the atmosphere. The heat exchange with the atmosphere moder-58 ates European climate (Rahmstorf, 2002) and transforms the water into cooler, denser 59 Subpolar Mode Water (Pérez-Brunius et al., 2004; McCartney & Talley, 1982; Brambilla 60 & Talley, 2008). This process of converting water parcels from one density class to an-61 other is referred to as water mass transformation (WMT). 62

AMOC exhibits substantial low-frequency variability in global climate models 63 (e.g., Kwon and Frankignoul (2014); Delworth and Zeng (2016); MacMartin et al. (2016)), 64 which has large effects on both North Atlantic and Arctic climate (e.g., Covey and Thomp-65 son (1989); Day et al. (2012); Zhang (2015); Oldenburg et al. (2018)). Low-frequency 66 AMOC variability is associated with variations in the upper-ocean density in the north-67 ern subpolar gyre (Roberts et al., 2013; Robson et al., 2016) as well as North Atlantic 68 sea-level pressure (SLP) patterns associated with changes in the North Atlantic Os-69 cillation (NAO; Eden and Jung (2001); Mecking et al. (2015); Delworth et al. (2016); 70 Delworth and Zeng (2016); Kim et al. (2018, 2020)). 71

AMOC is closely linked to the subpolar North Atlantic WMT (Marsh, 2000; Isach sen et al., 2007; Grist et al., 2009; Josey et al., 2009; Langehaug, Rhines, et al., 2012),
 which is responsible for driving high-latitude deep water formation. The link between
 WMT and AMOC has been the subject of many studies, mainly using low-resolution

-3-

 (~ 1) global climate models (e.g. (Langehaug, Rhines, et al., 2012)). However, low-resolution 76 global climate models differ in terms of which deep water formation regions domi-77 nate the AMOC structure and variability (e.g. Langehaug, Rhines, et al. (2012); Menary 78 et al. (2015); Heuze (2017); Oldenburg et al. (2021)). The biases in the deep water for-79 mation regions coincide with biases in subpolar temperature and salinity relative to 80 observations (Langehaug, Rhines, et al., 2012). In addition, Nordic Seas overflow pro-81 cesses, which are responsible for producing the dense water masses that make up the 82 southward flowing portion of AMOC and occur at relatively small spatial scales (Treguier 83 et al., 2005; Langehaug, Medhaug, et al., 2012), are too weak in many low-resolution 84 ocean models (Bailey et al., 2005). This results in a deficit in the volume transport of 85 these water masses. Moreover, low-resolution models do not resolve ocean mesoscale 86 eddies, which are known to contribute to water-mass transformation via convection 87 and lateral buoyancy fluxes, particularly in the Labrador Sea (Garcia-Quintana et al., 88 2019). 89

In low-resolution simulations, low-frequency AMOC variability appears to be 90 driven primarily by Labrador Sea WMT changes, regardless of where the climatolog-91 ical WMT is concentrated (Oldenburg et al., 2021). The mechanism of the low-frequency 92 AMOC variability involves upper ocean cooling and densification in the Labrador Sea, 93 driven by northwesterly winds off eastern North America. This increases deep con-94 vection there, which later strengthens AMOC and OHT. The strengthened AMOC and 95 OHT carry anomalous warm water northward into the subpolar regions, reducing deep 96 convection and AMOC and OHT. This mechanism, dominated by Labrador Sea WMT 97 variability, holds true across three low-resolution models with distinct representations 98 of deep water formation in subpolar regions (Oldenburg et al., 2021). However, one 99 concern with these results is that low-resolution simulations likely overestimate deep 100 water formation and subduction in the Labrador Sea region compared to high-resolution 101 ocean simulations (Garcia-Quintana et al., 2019). This is because of the large role that 102 convective eddies play during the restratification phase in the spring and summer months. 103 Mixed-layer depths are also likely too deep in low-resolution models owing to the ab-104 sence of eddies (Garcia-Quintana et al., 2019). This raises several interesting questions: 105 (1) Do the mechanisms of low-frequency AMOC and OHT variability found in low-106 resolution models, where the Labrador Sea appears to be the most important region 107 for initiating AMOC variability (Oldenburg et al., 2021), still hold in a high-resolution 108

-4-

model? (2) How does the ocean resolution of a model affect the partitioning of WMT
 between the different deep water formation regions?

In this paper, we aim to evaluate how well a high-resolution coupled model re-111 produces the surface-forced WMT found in a high-resolution atmospheric reanalysisforced ocean simulation, which we consider as an approximation to observations, and 113 compare that to what is found in a low-resolution version of the same model. We then 114 analyze the factors that set the magnitude of WMT in these simulations. Finally, we 115 examine the mechanisms of low-frequency AMOC variability in the high- and low-116 resolution versions of the coupled model. We focus in particular on the link between 117 the AMOC variability and the WMT variability in the different deep-water formation 118 regions and on how the variability is affected by the differences in resolution and mean 119 state. 120

In Section 2, we describe the model simulations used in this analysis. In Section 3, we compute the WMT and AMOC in the different simulations and analyze the factors that explain the differences between them. In Section 4, following the methods of Oldenburg et al. (2021), we use a low-frequency component analysis (LFCA) to elucidate the mechanisms of low-frequency AMOC variability in the high- and low-resolution versions of the coupled model. In Section 5, we summarize and discuss the overall results and conclusions.

128 **2** Description of models

We use output from a 1800-year pre-industrial control simulation of the Com-129 munity Earth System Model Version 1.1 (CESM1.1, Hurrell (2013)), with a nominal hor-130 izontal resolution of 1° in the atmosphere and ocean. We henceforth refer to this low-131 resolution CESM1 simulation as CESM1-LR. We also use output from a 500-year pre-132 industrial control simulation of CESM1.3 by the International Laboratory for High-133 Resolution Earth System Prediction (iHESP) (Chang et al., 2020), which uses an eddy-134 resolving 0.1° version of the Parallel Ocean Component version 2 (POP2) and a 0.25° 135 version of the Community Atmosphere Model version 5 (CAM5). We henceforth re-136 fer to this high-resolution CESM1 simulation as CESM1-HR. Unlike its low-resolution 137 counterpart, this model does not include a parameterization for overflows of deep wa-138 ter from the Nordic Seas into the North Atlantic while still not fully resolving the over-139

flow processes. Here we analyze the last 350 years of the 500-year simulation, because
the first 150 years are considered spin-up.

For our analysis of reanalysis-forced ocean-sea-ice simulations, we use output 142 from 1° and 0.1° POP2 ocean simulations, respectively, both forced with atmospheric 143 reanalysis data from the Japanese 55-year Reanalysis (JRA-55, Kobayashi et al. (2015); 144 Harada et al. (2016); Kim et al. (2021)), spanning years 1958-2018. Henceforth, we re-145 fer to these low- and high-resolution simulations as JRA55-LR and JRA55-HR, respec-146 tively. Here we are seeking an analogue to observations which still provides full ocean 147 output data. Given that historical ocean observations are limited to particular regions 148 or require reconstruction from proxies, an atmospheric reanalysis-forced ocean sim-149 ulation, which includes an ocean constrained at the surface to best estimates of his-150 torical atmospheric states, is a useful alternative. It would be possible to instead use 151 ocean assimilation data. However, they typically do not have closed heat and salt bud-152 gets, which are important when linking WMT to the interior ocean state. Also, his-153 torical ocean observations are fairly limited compared to atmospheric observational 154 data, which reduces the reliability of assimilation products. Hence, we take JRA55-155 HR as our closest analogue to observations. 156

Here we compare the rest of the simulations to JRA55-HR to determine whether increasing the ocean and atmospheric resolution of a coupled model leads to a more accurate representation of WMT and AMOC. Comparing JRA55-LR with CESM1-LR illustrates the role of atmospheric forcing (reanalysis data versus a coupled atmosphere) at the same ocean model resolution, while comparing JRA55-LR with JRA55-HR illustrates the role of ocean model resolution (parameterized versus resolved mesoscale eddies) under the same atmospheric forcing.

¹⁶⁴ 3 Comparison of WMT and AMOC climatologies

Before analyzing WMT and AMOC, it is helpful to consider the time-mean winter (January-February-March) mixed-layer depth to determine where the deep convection and deep water formation occur in the different models. In JRA55-HR, deep mixed layers are concentrated mostly in the Labrador Sea and Irminger and Iceland Basins (IIB), with some deep mixed layers in the Greenland-Iceland-Norwegian (GIN) Seas as well (Fig. 1a). In JRA55-LR, the mixed layers overall are deeper, and the deep-

-6-

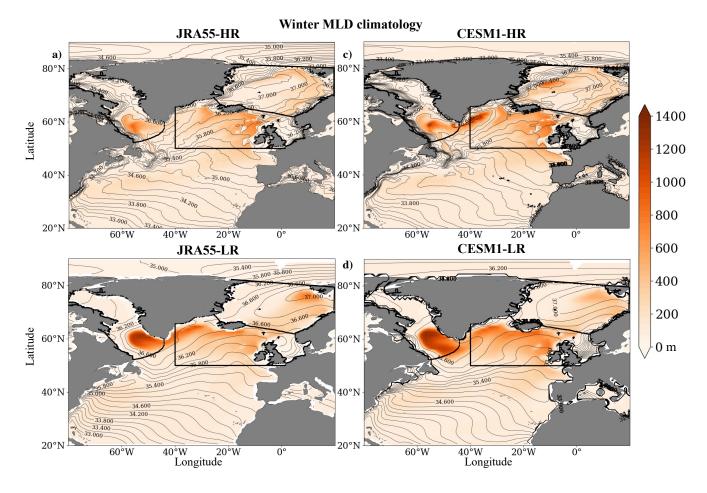


Figure 1: Climatological mixed-layer depth (colors) and sea-surface potential density referenced to 2000 m (contours) both averaged over January, February and March in **a**) JRA55-HR, **b**) JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR. The thick black lines represent the region masks for the Labrador Sea (left), Irminger-Iceland Basins (lower right) and GIN Seas (upper right).

est mixed layers are concentrated in the Labrador Sea, though there are still deep mixed 171 layers in the IIB and GIN Seas (Fig. 1b). In CESM1-HR, the mixed-layer depth pat-172 terns look similar to JRA55-HR, but the mixed-layer depths are deeper in all of the 173 deep water formation regions (Fig. 1c). In CESM1-LR, the deepest mixed layers are 174 mostly concentrated in the Labrador Sea, even more so than in JRA55-LR, which shows 175 similar overall patterns (Fig. 1b, d). It is noteworthy that CESM1-HR captures the mixed-176 layer depth patterns found in JRA55-HR much better than either of the low-resolution 177 models, despite JRA55-LR being forced with the same atmospheric reanalysis data as 178 JRA55-HR. 179

Throughout our analysis, we use AMOC calculated in density coordinates, rather than AMOC calculated in depth coordinates, because it is more appropriate for analyzing subpolar AMOC variability and is strongly connected to the the analysis of WMT as a function of density class (Straneo, 2006; Pickart & Spall, 2007). We first look at the AMOC climatology to determine how well the coupled simulations (and JRA55-LR) reproduce the AMOC from the reanalysis-forced high-resolution dataset, JRA55-HR. To compute AMOC, we use Eq. (1) from Newsom et al. (2016):

$$AMOC(\sigma, y, t) = -\int_{x_W}^{x_E} \int_{-B(x,y)}^{z(x,y,\sigma,t)} v(x,y,z,t) dz dx,$$
(1)

where σ is the potential density referenced to 2000m, y is the latitude, x is longitude, x_W and x_E are the western and eastern longitudinal limits of the basin, respectively, v is the meridional velocity, z is depth (positive upwards), B(x, y) is the bottom depth, and t is time.

In JRA55-HR, the maximum AMOC is located at $\sigma_2 = 36.48$ kg m⁻³, where it 191 reaches 21.8 Sv (Fig. 2a). In JRA55-LR, the maximum is located at $\sigma_2 = 36.58$ kg m⁻³ 192 and is 20.7 Sv (Fig. 2b). AMOC in CESM1-HR reaches a maximum of 25.4 Sv at $\sigma_2 =$ 193 36.53 kg m⁻³ (Fig. 2c). In CESM1-LR, AMOC reaches a maximum of 28.6 Sv at $\sigma_2 =$ 194 36.64 kg m⁻³ (Fig. 2d). Hence, in terms of maximum magnitude, JRA55-LR reproduces 195 the AMOC found in JRA55-HR the best of all the other model simulations, though 196 CESM1-HR reproduces the density where the maximum occurs most accurately. Sur-197 prisingly, the maximum AMOC is actually smaller in JRA55-LR than in JRA55-HR; 198 we would expect a higher resolution simulation to yield a weaker AMOC, as in CESM1-199 HR and CESM1-LR, and also what was found in other studies of coupled GCMs (Winton, 200

-8-

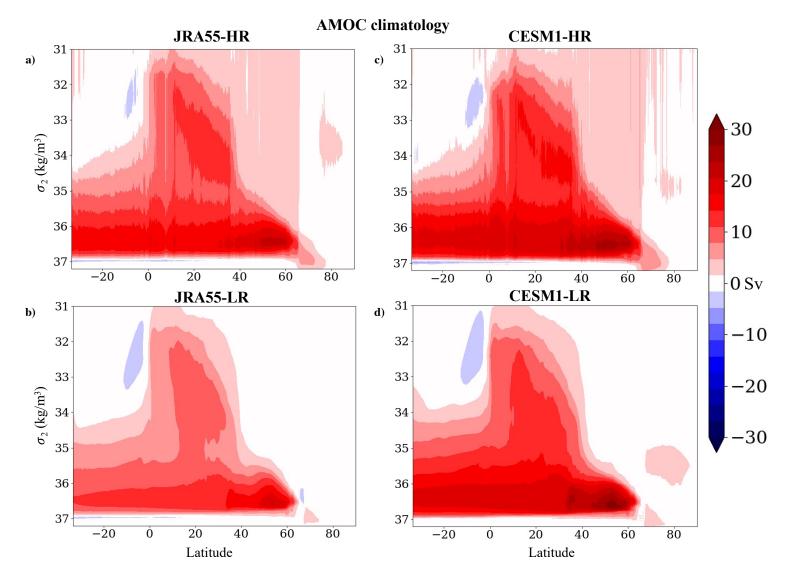


Figure 2: Climatological AMOC in **a**) JRA55-HR, **b**) JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

201 2014; Sein et al., 2018). All of the simulations have AMOC maxima located at higher 202 densities than JRA55-HR. CESM1-HR has a maximum AMOC at a density closest to 203 the JRA55-HR maximum, while CESM1-LR has a maximum AMOC at a density fur-204 thest from the JRA55-HR maximum. These results indicate that although increasing 205 the resolution of the atmosphere and ocean yields an AMOC substantially closer to 206 reanalysis-forced ocean data, there are likely biases in the atmospheric component of 207 the coupled simulations even at high resolution.

To compute the surface-forced WMT, we use the equations described in Speer and Tziperman (1992) and also used in many other studies such as Langehaug, Rhines, et al. (2012). We first calculate the surface density flux D(x, y, t) using air-sea heat and freshwater fluxes (Walin, 1982; Tziperman, 1986; Speer & Tziperman, 1992):

$$D(x, y, t) = \frac{\alpha(x, y, t)Q_H(x, y, t)}{c_w} - \beta(x, y, t)S(x, y, t)Q_F(x, y, t),$$
(2)

The first and second terms here are the thermal and haline components, respectively, 208 computed in units of kg m⁻² s⁻¹. $\alpha(x, y, t)$ here is the thermal expansion coefficient 209 calculated at each grid point for every month, Q_H is the surface heat flux into the ocean 210 in W m⁻²; c_w is the specific heat capacity of seawater, assumed to be constant and uni-211 form and equal to 4186 J kg⁻¹ K⁻¹; $\beta(x, y, t)$ is the haline contraction coefficient, also 212 computed for each month at each grid point; S is the sea-surface absolute salinity; and 213 Q_F is the freshwater flux in units of kg m⁻² s⁻¹. The surface heat flux used here in-214 cludes fluxes of net shortwave and longwave radiation, heat fluxes due to sea-ice changes, 215 and latent and sensible heat fluxes. The freshwater flux is equal to the sum of the evap-216 oration, runoff, precipitation, sea-ice melt and formation fluxes. All of these variables 217 are from monthly model output model. 218

We integrate this density flux, D(x, y, t), over all grid boxes for each density class to calculate the surface-forced WMT:

$$F(\sigma) = \frac{1}{\Delta\sigma} \int_{\sigma}^{\sigma + \Delta\sigma} D(x, y, t) dA,$$
(3)

Here, $F(\sigma)$ refers to the surface-forced WMT in units of Sv, $\sigma = \rho - 1000$ is the potential density in units of kg m⁻³ referenced to 2000m, and $\Delta \sigma$ is the density bin width. Here, as in Oldenburg et al. (2021), we neglect the mixing contributions because the model output data do not have sufficient time resolution to calculate them. We compute the WMT separately in the Labrador Sea, Irminger and Iceland Basins (IIB) and

-10-

GIN Seas using the region masks shown in the boxes in Fig. 1 to determine each region's contribution to the total WMT.

In all four simulations, the thermal WMT component dominates over the haline 226 contribution. However, the partitioning of WMT in the different regions varies sub-227 stantially among the simulations. In JRA55-HR, none of the peaks in WMT in the dif-228 ferent regions align with the density of maximum AMOC. The IIB contributes the most 229 to the WMT at densities lower than the density of maximum AMOC (Fig. 3a), reach-230 ing a maximum value of 14.2 Sv at $\sigma_2 = 36 \text{ kg/m}^3$. At densities higher than the max-231 imum AMOC, the WMT is dominated by contributions from the Labrador Sea and 232 GIN Seas, with a much narrower peak in the Labrador Sea. The Labrador Sea has a 233 peak of 7.7 Sv at $\sigma_2 = 36.7 \text{ kg/m}^3$, and the GIN Seas WMT peaks at 4.6 Sv at $\sigma_2 =$ 234 36.56 kg/m³. Though these densities are further away from the maximum AMOC, 235 they are likely still important for AMOC given that internal mixing acts to reduce the 236 density of the densest water masses. In JRA55-LR, the peaks in the IIB and GIN Seas 237 WMT occur closer to the maximum AMOC, reaching maxima equal to 14.5 and 6.2 238 Sv at $\sigma_2 = 36.32$ and $\sigma_2 = 36.62$, respectively, and the IIB dominates the WMT near 239 the AMOC maximum (Fig. 3b). The Labrador Sea peak in WMT is located at about 240 the same density as in JRA55-HR, with a peak value of 11.4 Sv at $\sigma_2 = 36.7 \text{ kg/m}^3$. 241 Furthermore, the peaks in the Labrador Sea and GIN Seas WMT are narrower in JRA55-242 LR than they are in JRA55-HR. 243

The WMT in CESM1-HR looks the most similar to JRA55-HR of all the other sim-244 ulations, with the most notable difference being that the WMT peaks in the IIB and 245 Labrador Sea WMT are larger than in JRA55-HR (Fig. 3c), with the IIB WMT reach-246 ing a maximum value of 17.4 Sv at $\sigma_2 = 36 \text{ kg/m}^3$, the Labrador Sea WMT reach-247 ing a maximum of 8.3 Sv at $\sigma_2 = 36.74 \text{ kg/m}^3$, and the GIN Seas WMT peaking at 248 5.0 Sv at $\sigma_2 = 36.74 \text{ kg/m}^3$. However, the partitioning of the WMT between the dif-249 ferent regions remains similar to JRA55-HR. In CESM1-LR, on the other hand, the WMT 250 looks quite different, with much larger WMT peaks in the IIB and the Labrador Sea 251 WMT than in any of the other simulations (Fig. 3d), reaching maxima equal to 19.6 252 and 21.2 Sv at $\sigma_2 = 36.26$ and $\sigma_2 = 36.72$, respectively. The peak in Labrador Sea 253 WMT is also much narrower than in JRA55-HR and CESM1-HR, and looks more sim-254 ilar to JRA55-LR. The GIN Seas WMT peaks at $\sigma_2 = 36.82 \text{ kg/m}^3$, where it reaches 255 a maximum of 6.9 Sv. This seems to indicate that increasing the atmospheric and ocean 256

-11-

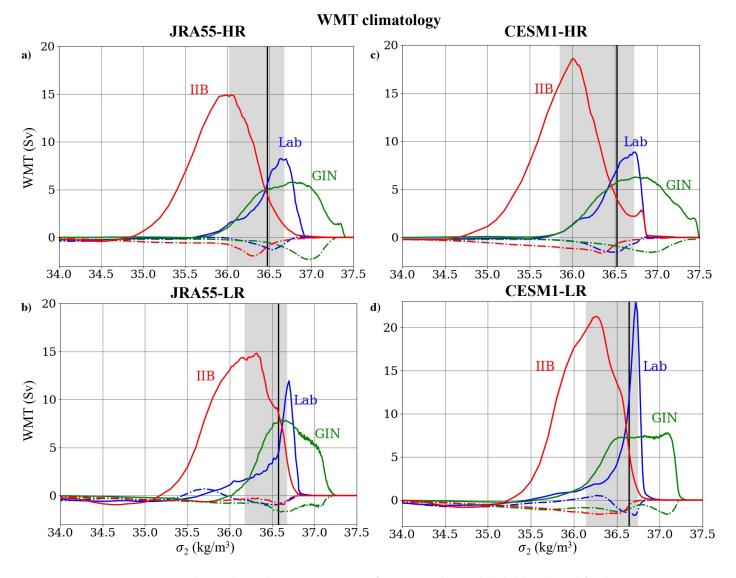


Figure 3: Climatological water-mass transformation thermal (solid lines) and freshwater (dashed lines) components in the Labrador Sea (Lab), GIN Seas and Irminger and Iceland Basins (IIB) for **a**) JRA55-HR, **b**) JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR. The black vertical lines indicate the density where the climatological AMOC reaches its maximum in each model. The grey shaded areas represent the density range where AMOC is within 25% of its maximum value. A more detailed illustration of what particular areas of the deep water formation regions contribute to the surface density flux over different density classes is shown in Figures 4-5, as well as Fig. S1.

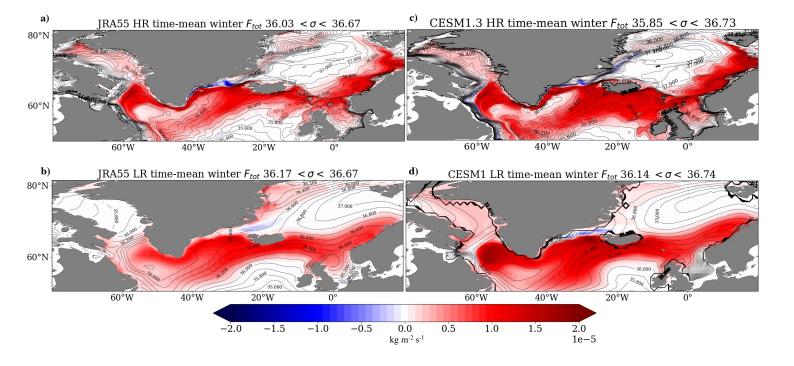


Figure 4: Colors: Total climatological winter surface density flux D(x, y, t), calculated using Eq. (2) over densities where AMOC is at least 75% of its maximum. Contours: Time-mean winter sea-surface potential density referenced to 2000 m for **a** JRA55-HR, **b** JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

resolution in a coupled model yields a fairly realistic representation of WMT in the
different deep water formation regions, certainly much more realistic than an equivalent low-resolution coupled model. The major discrepancies between JRA55-LR and
JRA55-HR indicate that a higher ocean model resolution is essential in order to provide an accurate representation of WMT; having correct atmospheric surface forcing
alone is insufficient.

To illustrate which parts of each region contribute to the WMT in different density classes, it is useful to look at the full surface-density flux D(x, y, t) calculated from Eq. (2). Since we are interested in the density classes relevant for AMOC, we isolate the D(x, y, t) for densities lower than the minimum density where AMOC reaches 75% of its maximum (Fig. S1), densities within the density range where AMOC is at least

-13-

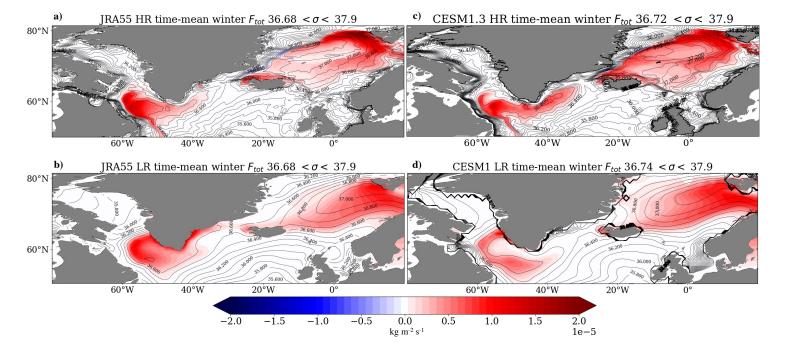


Figure 5: Colors: Total climatological winter surface density flux D(x, y, t), calculated using Eq. (2) over densities above the maximum density where AMOC reaches 75% of its maximum. Contours: Time-mean winter sea-surface potential density referenced to 2000 m for **a** JRA55-HR, **b** JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

75% of its maximum (Fig. 4), and densities above that density range (Fig. 5). In the lowest density range, the surface-density flux is concentrated in the Irminger and Iceland Basins, with small contributions from the other regions, mainly near coastlines where the water is fresher and lighter than the interior areas (Fig. S1). Because interior mixing tends to reduce the density of water parcels, the surface-density fluxes in this density range are unlikely to contribute to AMOC.

In the density range near the AMOC maximum, CESM1-HR reproduces the den-274 sity flux patterns found in JRA55-HR fairly well. In both of these simulations, most 275 of the Labrador Sea surface density flux is concentrated in the northern section of the 276 Labrador Sea rather than in the southern section, where density fluxes are weaker (Fig. 277 4a, c). The patterns found in the GIN Seas are also similar; however, the surface den-278 sity fluxes in the southern part of the IIB are much higher in CESM1-HR than in JRA55-279 HR (Fig. 4a, c). The low-resolution simulations show similar overall patterns to JRA55-280 HR, but lack several key features (Fig. 4b, d). For example, Labrador Sea fluxes are 281 more concentrated in the central and southern sections compared to JRA55-HR and 282 CESM1-HR, particularly in CESM1-LR (Fig. 4d). JRA55-LR reproduces the flux pat-283 terns in the IIB fairly well (Fig. 4b). However, neither low-resolution simulation has 284 an accurate representation of the more complex smaller scale density structures found 285 in JRA55-HR and CESM1-HR, where the densities are less uniform, particularly near 286 coastlines. For the highest density range, the interior and southern parts of the Labrador 287 Sea contribute more to WMT in JRA55-HR and CESM1-HR compared to the lower den-288 sity classes (Fig. 5a, c). There are also larger contributions from the interior and north-289 ern parts of the GIN Seas. The same overall patterns are found in the low-resolution 290 simulations (Fig. 5b, d). However, in JRA55-LR the surface density fluxes in the Labrador 291 Sea are shifted to the east relative to JRA55-HR and CESM1-HR, and the northern part 292 of the GIN Seas is not emphasized as much as in the high-resolution simulations, with 293 a much more uniform pattern in the eastern GIN Seas (Fig. 5b). In CESM1-LR, the con-294 tributions to WMT from the Labrador Sea are smaller, and the eastern area of the GIN 295 Seas is more emphasized compared to in JRA55-LR (Fig. 5d). 296

To allow for a more direct comparison between AMOC and the WMT in the different regions, we also calculate the surface-forced overturning streamfunction follow-

-15-

ing the methodology of Marsh (2000):

$$F(\sigma,\Theta,t) = -\frac{\partial}{\partial\sigma} \int \int_{\theta>\Theta,\sigma^*>\sigma} D(x,y,t) dA,$$
(4)

where Θ is the latitude; θ is a dummy variable representing the latitude; σ is the seasurface density referenced to 2000m; σ^* is a dummy variable representing the sea-surface density; D(x, y, t) is the density flux calculated in Eq. 2; *t* is the time; and *A* is the surface area.

Here we calculate the surface-forced overturning streamfunction for each of the 301 three regions separately, which allows us to quantify how much the surface-forced WMT 302 in each region contributes to AMOC (neglecting mixing). CESM1-HR reproduces the 303 surface-forced overturning found in JRA55-HR far better than either low-resolution 304 simulation in all regions (Fig. 6a-d, i-l). In JRA55-LR and CESM1-LR, the overturn-305 ing is too strong in all the regions, especially in the Labrador Sea and IIB (Fig. 6e-h, 306 m-p). Also, the Labrador Sea surface-forced overturning is concentrated over a smaller 307 density range in the LR models compared to the HR versions (Fig. 6b, f, j, n). For the 308 IIB, overturning in the HR simulations is shifted towards lower densities compared 309 to the LR versions (Fig. 6c, g, k, o). Overturning in the GIN Seas is also concentrated 310 over a smaller density range in the LR models than in the HR models (Fig. 6d, h, l, 311 p). 312

To determine what is responsible for the discrepancies in the WMT between JRA55-313 HR and the other simulations, we discuss the climatologies of several surface prop-314 erties used in the WMT calculation, including the sea-surface heat fluxes as well as 315 the sea-surface potential temperatures, salinities and densities. Although the fresh-316 water fluxes also contribute to the WMT, the freshwater components of WMT are very 317 small in all four simulations (Fig. 3). Hence we do not show them here, but rather in 318 the supplementary section (Fig. S2). For these quantities, we present the climatology 319 in JRA55-HR (Fig. 7e) and the anomalies for the other simulations relative to JRA55-320 HR. CESM1-HR shows a much more accurate representation of the time-mean den-321 sity structure compared to both low-resolution simulations, particularly in the Labrador 322 Sea and near all coastlines (Fig. 7f). CESM1-HR anomalies in sea-surface temperatures 323 and salinities relative to JRA55-HR are more substantial than its density anomalies (Fig. 324 8b, f), but they are mostly density compensating, yielding smaller density anomalies. 325 These anomalies lead to small positive density anomalies in the GIN Seas, IIB and Labrador 326

-16-

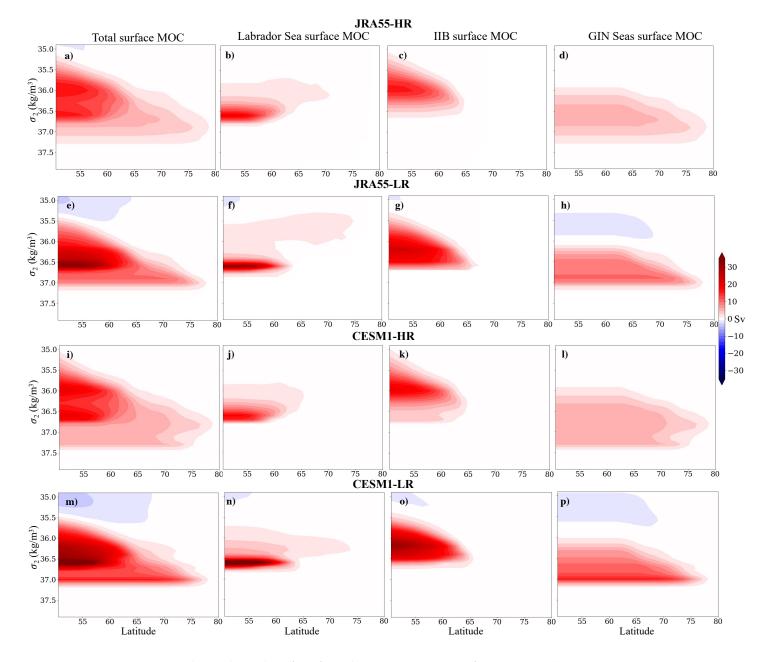


Figure 6: Climatological surface-forced overturning streamfunction in **a-d**) JRA55-HR, **e-h**) JRA55-LR, **i-l**) CESM1-HR and **m-p**) CESM1-LR computed over all regions (first column), the Labrador Sea (second column), the Irminger-Iceland Basins (IIB, third column) and GIN Seas (fourth column).

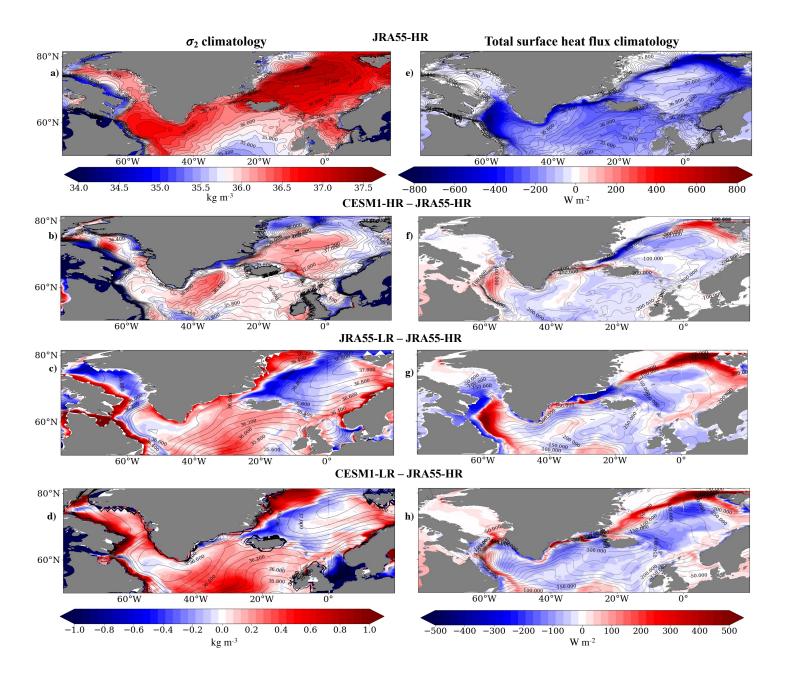


Figure 7: **a**) JRA55-HR climatology of sea-surface potential density, referenced to 2000m. **b-d**) Sea-surface potential density climatologies (contours) and anomalies relative to JRA55-HR (colors) for **b**) CESM1-HR, **c**) JRA55-LR and **d**) CESM1-LR. **e**) JRA55-HR total sea-surface heat flux climatology. **f-h**) Sea-surface heat flux climatologies (contours) and anomalies relative to JRA55-HR (colors) for **f**) CESM1-HR, **g**) JRA55-LR and **h**) CESM1-LR.

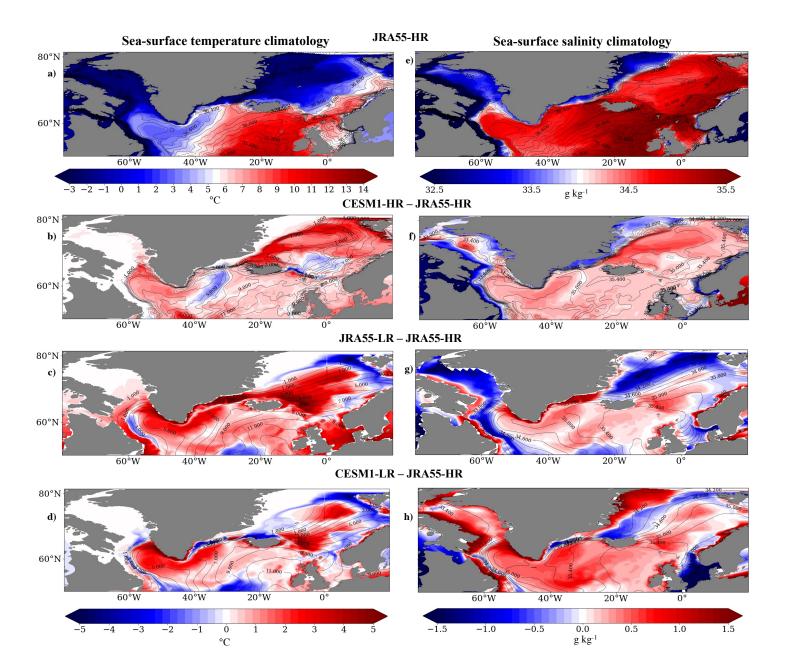


Figure 8: **a**) JRA55-HR sea-surface potential temperature climatology. **b-d**) Sea-surface potential temperature climatologies (contours) and anomalies relative to JRA55-HR (colors) for **b**) CESM1-HR, **c**) JRA55-LR and **d**) CESM1-LR. **e**) JRA55-HR sea-surface salinity climatology. **f-h**) Sea-surface salinity climatologies (contours) and anomalies relative to JRA55-HR (colors) for **f**) CESM1-HR, **g**) JRA55-LR and **h**) CESM1-LR.

Sea, except near the coastlines (Fig. 7f), likely due to increased freshwater runoff com-327 pared to JRA55-HR (Fig. S2). JRA55-LR, on the other hand, shows large negative den-328 sity anomalies in the central GIN Seas, but positive anomalies near the coastlines (Fig. 329 7g). There are also positive anomalies in the eastern subpolar gyre and in the north-330 ern Labrador Sea. The density structure looks similar in CESM1-LR, with similar anoma-331 lies relative to JRA55-HR in most regions, except for in the northern Labrador Sea where 332 there are actually positive anomalies (Fig. 7h), due to a fairly salty Labrador Sea com-333 pared to the other simulations (Fig. 8h). The higher densities in the low-resolution sim-334 ulations explain why the WMT and AMOC peaks occur at higher densities than in 335 JRA55-HR and CESM1-HR (Fig. 3), and the generally more uniform density fields in 336 the Labrador Sea explain the narrower WMT peaks in the LR simulations compared 337 to JRA55-HR and CESM1-HR. Also, the high densities in the GIN Seas in CESM1-HR 338 explain why there is positive WMT in that region at higher densities than what is seen 339 in the other models (Fig. 3c). 340

CESM1-HR best reproduces the surface heat fluxes found in JRA55-HR (Fig. 7a, 341 b), with some positive anomalies in the central and northern Labrador Sea and broad 342 negative anomalies throughout the IIB and GIN Seas, aside from the far north, which 343 exhibits positive anomalies (Fig. 7b). The larger (more negative) heat fluxes in the IIB 344 and GIN Seas explain the larger IIB and GIN WMT in CESM1-HR compared to JRA55-345 HR, given that stronger heat fluxes drive higher WMT. JRA55-LR exhibits larger pos-346 itive anomalies in the Labrador Sea and northern GIN Seas compared to CESM1-HR 347 (Fig. 7c). In CESM1-LR, there is a mix of positive and negative anomalies in the Labrador 348 Sea, and larger negative anomalies in the central GIN Seas (Fig. 7d). 349

Surprisingly, CESM1-HR reproduces the WMT, sea-surface heat fluxes, sea-surface temperatures and salinities of JRA55-HR far better than JRA55-LR does, which highlights the importance of ocean resolution in accurately representing these variables. It also indicates that simply forcing an ocean model with atmospheric reanalyses is insufficient if the ocean is low-resolution.

4 Mechanisms of low-frequency AMOC variability in high- and low-resolution versions of CESM

³⁵⁷ We next turn our attention to the mechanisms driving low-frequency AMOC vari-³⁵⁸ ability. Following the methods of Oldenburg et al. (2021), we apply a low-frequency

-20-

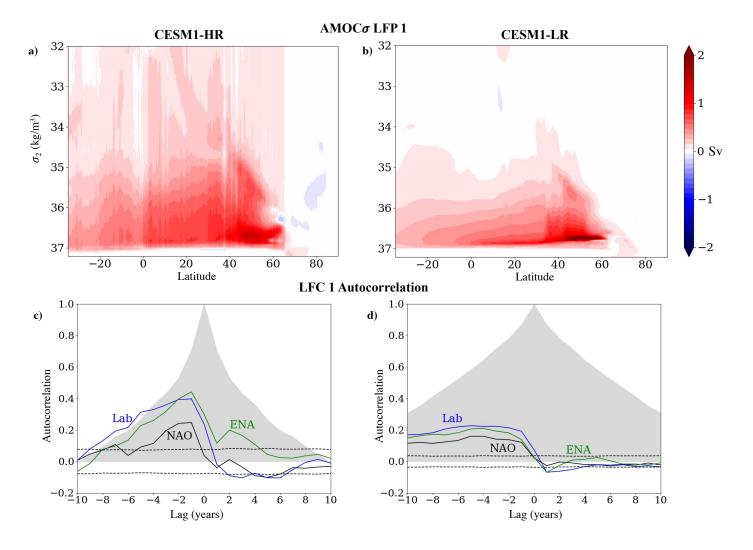


Figure 9: Top row: LFP 1 of AMOC for **a**) CESM1-HR and **b**) CESM1-LR. Bottom row: Autocorrelations of LFC 1 (shaded), correlation of the NAO with LFC 1 (solid black lines) and significance levels (dashed black lines), and correlations of both the Labrador Sea (blue lines) and Eastern North Atlantic (ENA; green lines) winter mixed-layer depths with LFC 1 for **c**) CESM1-HR and **d**) CESM1-LR. NAO here is defined as the difference between the sea-level pressure between the Azores (25.5°W, 37.5°N) and Iceland (21.5°W, 64.5°N). The ENA here includes both the Irminger and Iceland Basins and the GIN Seas.

component analysis (LFCA; R. C. Wills et al. (2018); R. C. J. Wills et al. (2019)) to AMOC 359 in density coordinates in CESM1-HR and CESM1-LR's pre-industrial control simula-360 tions. We find the low-frequency patterns (LFPs) of AMOC, which are the linear com-361 binations of the leading empirical orthogonal functions (EOFs) that maximize the ra-362 tio of low-frequency variance to total variance in their corresponding timeseries (called 363 low-frequency components; LFCs). Low-frequency variance is defined as the variance 364 that remains after the point-wise application of a Lanczos filter with a low-pass cut-365 off of 10 years. The 10-year low-pass filter is only used in identifying the LFPs, and 366 all information about high-frequency variations in the data is preserved. We focus on 367 the first LFP/LFC (Fig. 9), which has the highest ratio of low-frequency variance to 368 total variance and is well separated in this ratio from the second LFP/LFC. This LFP 369 represents the AMOC anomaly associated with a one standard deviation (1σ) anomaly 370 in the corresponding LFC time series. For both models, when calculating the LFPs/LFCs, 371 we include the six leading EOFs. The choice of the number of EOFs does not substan-372 tially change the results for any of the models. 373

In our previous analysis of low-resolution coupled model simulations (Oldenburg 374 et al., 2021), we found that WMT in the Labrador Sea plays a more substantial role 375 in driving AMOC and OHT variability than would be expected based on its role in 376 driving the climatology of AMOC and OHT. Here, we examine whether the model 377 resolution affects this result, given that higher resolution models represent Labrador 378 Sea processes much better than low-resolution ones (see section 3). Hence, here we 379 carry out an analysis similar to Oldenburg et al. (2021) with a focus entirely on AMOC 380 instead of Atlantic OHT. Our goal is to determine whether the mechanisms of low-381 frequency AMOC variability in low-resolution simulations still hold in high-resolution 382 models. We first compute the LFPs and LFCs of annual-mean AMOC in CESM1-HR 383 and CESM1-LR, then calculate lead-lag regressions between the first LFC and other 384 fields, including winter mixed-layer depth, surface-forced WMT, winter sea-level pres-385 sure (SLP) and AMOC. Although the LFPs already give the AMOC anomaly at lag-386 0, the pattern of AMOC anomalies evolves over time and therefore can look differ-387 ent at lead and lag times. 388

The first LFPs of AMOC in CESM1-HR and CESM1-LR share some common features, with maxima in the mid to subpolar latitudes. In CESM1-HR, the maximum value is equal to 1.48 Sv and is located at 47° N and $\sigma_2 = 36.675 \text{ kg/m}^3$. In CESM1-LR,

-22-

the maximum value is equal to 2.51 Sv and is located at 53.5° N and $\sigma_2 = 36.74 \text{ kg/m}^3$. 392 This is substantially stronger and at a higher latitude and density than in CESM1-HR. 393 The peak is also broader in CESM1-HR. The other major difference is that the pos-394 itive values extend to lower densities in CESM1-HR compared to CESM1-LR. The ra-395 tios of low-frequency to total variance for the LFPs are equal to 0.70 and 0.87 for CESM1-396 HR and CESM1-LR, respectively. The LFC autocorrelations remain high for much longer 397 lag times in CESM1-LR compared to CESM1-HR (Fig. 9c, d). In CESM1-HR, the au-398 tocorrelation drops off more quickly, reaching zero by lag 10 years (Fig. 9c). The lower 399 ratio of low-frequency to total variance in CESM1-HR indicates that that model's LFC 400 includes more high-frequency variability, and the lower autocorrelation is consistent 401 with an AMOC that changes more rapidly over lead and lag times (Fig. S3). 402

In CESM1-HR, there is a persistent SLP pattern associated with anomalous north-403 westerly winds off eastern North America starting about four years before the time 404 of maximum AMOC (Fig. 10b). This pattern remains until lag zero, which is the time 405 of maximum AMOC (Fig. 10a-d). Because the persistence time scale of SLP anoma-406 lies is less than one month (Ambaum & Hoskins, 2002), persistence of this pattern must 407 be due to memory coming from the ocean. At lag zero, the SLP pattern becomes more 408 zonal and the eastern SLP intensifies (Fig. 10d. After lag zero, the pattern reverses (Fig. 409 10e, f) with a pattern that looks similar to the negative phase of the NAO. In CESM1-410 LR, there is a similar SLP pattern at lead times and at lag-0 (Fig. 10g-j). In both HR 411 and LR models, the effect of the SLP pattern at lead times on the subpolar winter mixed-412 layer depths can be seen in Fig. S4, which shows deepening mixed-layer depths, par-413 ticularly in the Labrador Sea. The time evolution of Labrador Sea mixed-layer depth 414 mirrors that of the NAO (Fig. 9c, d). The ENA mixed-layer depth does follow the NAO 415 to some degree, especially in CESM1-LR, but it doesn't mirror it to the same degree 416 as the Labrador Sea in CESM1-HR (Fig. 9c, d). After lag zero, the SLP pattern dissi-417 pates completely in CESM1-LR (Fig. 10l). However, unlike many low-resolution mod-418 els, including CESM1-LR and the LR models discussed in Oldenburg et al. (2021), CESM1-419 HR shows a coherent SLP pattern after the time of maximum AMOC. This indicates 420 an atmospheric response to the low-frequency AMOC variability not seen in the equiv-421 alent low-resolution model. This response can also be seen in the negative lagged cor-422 relation of the NAO with LFC 1 (Fig. 9c), which peaks at a lag of 5 years. 423

-23-



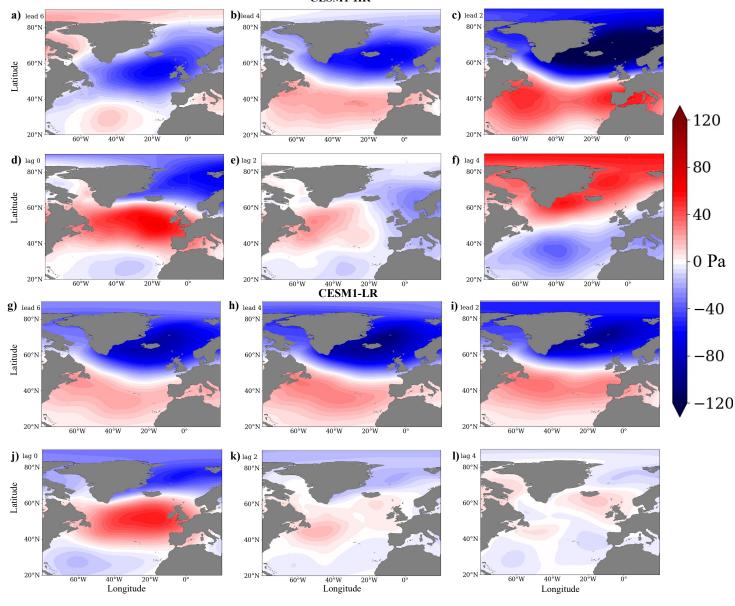


Figure 10: Lead-lag regressions of sea-level pressure averaged over January, February and March onto the first LFC of AMOC for **(a-f)** CESM1-HR and **(g-l)** CESM1-LR. Lead times indicate anomalies that lead the LFC, i.e., prior to the maximum AMOC. Because the LFCs are unitless, the regressions simply have units of Pa (N/m^2) .

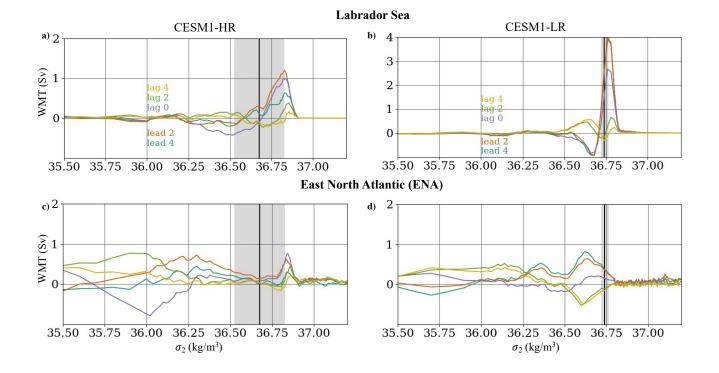


Figure 11: Lead-lag regressions of water mass transformation (WMT) onto the first LFC of AMOC for CESM1-HR (left column) and CESM1-LR (right column). **a**, **b**) WMT summed over the Labrador Sea region. **c**, **d**) WMT summed over the Eastern North Atlantic (ENA) section. The black vertical lines indicate the density where the AMOC regression at lag zero reaches its maximum in each model. The grey shaded areas represent the density range where the AMOC regression at lag zero is within 25% of its maximum value. The black lines in Fig. 1 show what we consider to be the Labrador Sea, the Irminger and Iceland Basins, and the GIN Seas. Lead means LFC 1 lags, i.e., prior to the maximum AMOC. Because the LFCs are unitless, the regressions simply have units of Sv.

The AMOC changes before and after the time of maximum AMOC can be seen 424 in Fig. S3, which shows a strengthening of AMOC at lead times and a weakening at 425 lag times in both models. In CESM1-HR, WMT in the Labrador Sea strengthens in the 426 years leading up to maximum AMOC, reaching its maximum at lead 2, concurrent 427 with the strengthening of AMOC and the deepening of mixed layers in the Labrador 428 Sea, IIB and GIN Seas (Fig. 11a). This peak is equal to 1.29 Sv and is located at $\sigma_2 =$ 429 36.83 kg/m³, which is at a substantially higher density than the location of the max-430 imum AMOC anomaly at lag zero, but is still within the density range of the broad 431 positive AMOC anomaly. After lead 2, the WMT rapidly decreases. The Eastern North 432 Atlantic (ENA) WMT, which includes both the GIN Seas and IIB, also increases at lead 433 times, peaking at lead one years (Fig. 11c). This peak is equal to 0.83 Sv and is located 434 at $\sigma_2 = 36.84$, which is further from the peak in AMOC than the Labrador Sea WMT 435 peak. The peak in ENA WMT is mostly due to changes in IIB WMT rather than the 436 GIN Seas (not shown). 437

In CESM1-LR, the Labrador Sea WMT also increases at lead times, reaching its maximum at lead 2 years (Fig. 11b). This maximum is equal to 3.99 Sv and is located at $\sigma_2 = 36.76 \text{ kg/m}^3$, which is at a slightly higher density than the maximum AMOC anomaly. The ENA WMT also strengthens at lead times, but already peaks by lead 4 years (Fig. 11d). This peak is equal to 0.82 Sv and is located at $\sigma_2 = 36.62 \text{ kg/m}^3$, which is at a substantially lower density than the maximum AMOC anomaly. This WMT increase is mostly due to changes in the IIB rather than the GIN Seas (not shown).

Based on these results, it appears that the mechanisms of AMOC variability between CESM1-HR and CESM1-LR are qualitatively similar but still have quantitative differences. In both models, the Labrador Sea plays a dominant role in driving lowfrequency AMOC variability, and the leading sea-level pressure patterns are similar. The primary differences are that CESM1-HR, unlike CESM1-LR, shows a substantial atmospheric response after the time of maximum AMOC, and that the Labrador Sea does not dominate the WMT variability as much as it does in CESM1-LR.

452 **5** Discussion and Conclusions

Based on the results from Section 3, a coupled model with increased atmospheric
 and ocean resolutions accurately reproduces the WMT, sea-surface temperatures and

-26-

455 sea-surface salinities found in a reanalysis-forced high-resolution ocean simulation.

The ocean resolution appears to be particularly important, as even a low-resolution ocean simulation forced with atmospheric reanalysis data doesn't represent the WMT as accurately as the high-resolution coupled model simulation. This illustrates the importance of resolving, rather than parameterizing, mesoscale eddies for the ability to accurately represent mixed-layer depth and deep water formation, particularly in the Labrador Sea.

The better representation of WMT is explained by a more accurate representation of the density structure in the high-resolution simulation compared to the lowresolution simulations, which have relatively uniform density fields in comparison, particularly in the Labrador Sea. Smaller discrepancies in surface heat fluxes in the deep water formation regions in the high-resolution simulation also help explain why it captures the climatological WMT better than the low-resolution simulations.

In section 4, we used LFCA to assess the mechanisms of low-frequency AMOC 468 variability in high- and low-resolution versions of the same model, finding that the 469 mechanisms are qualitatively similar but quantitatively different. The Labrador Sea 470 WMT still plays a major role in the WMT and AMOC variability in the high-resolution 471 model despite the fact that it shows a smaller role for the Labrador Sea in climato-472 logical WMT and AMOC than the low-resolution version. The analysis here neglects 473 interior ocean mixing. However, despite the fact that most of the Labrador Sea WMT 474 changes occur at higher densities than the AMOC changes, the Labrador Sea's dom-475 inance in AMOC variability likely still holds because mixing tends to make the dens-476 est water lighter. 477

One noteworthy difference between the simulations is that the high-resolution 478 model shows a substantial atmospheric response to the AMOC variability not seen 479 in the low-resolution version. This type of atmospheric response has been seen in a 480 study of a medium-resolution coupled model, but with a longer lag time between the 481 AMOC change and the negative NAO response (Wen et al., 2016). NAO-like responses 482 of differing signs to AMOC variability have also been found in other studies (Dong 483 & Sutton, 2003; Gastineau & Frankignoul, 2012; Gastineau et al., 2013; Frankignoul et 484 al., 2013, 2015). The model simulations we analyzed here do not give insight into whether 485 the atmospheric or oceanic resolution is responsible for the increased atmospheric re-486

-27-

sponse to AMOC variability in CESM1-HR, but recent work suggests that the atmospheric response to near-surface ocean anomalies is larger at higher atmospheric resolution (e.g., Czaja et al. (2019)). Overall, it appears that the mode of AMOC variability in the high-resolution model is associated with stronger anomalies in atmospheric
fields (i.e., sea-level pressure), while the low-resolution version is associated with stronger
anomalies in ocean fields, namely in the water-mass transformation, particularly in
the Labrador Sea.

Our results suggest that increasing the ocean and atmospheric resolution of a coupled model substantially improves the representation of climatological AMOC and WMT. However, the mechanisms driving low-frequency AMOC variability remain qualitatively similar even though the climatologies differ. This is consistent with what was found in three low-resolution coupled models with distinct representations of WMT in the different subpolar North Atlantic deep water formation regions, which all showed similar mechanisms of AMOC and OHT variability, with the Labrador Sea playing a dominant role (Oldenburg et al., 2021).

502 Acknowledgments

The authors are grateful for support from the National Science Foundation through grants OCE-1523641 and OCE-1850900 (D. O. and K. C. A.); and AGS-1929775 (R. C. J. W.). L. T. acknowledges support from NASA Ocean Surface Topography Science Team grant NNX17AH56G. We thank the CMIP5 climate modeling groups and iHESP for making their model output available. MATLAB and Python code for LFCA is available at https://github.com/rcjwills/lfca.

509 References

- Ambaum, M. H. P., & Hoskins, B. J. (2002). The NAO Troposphere-Stratosphere
 Connection. *Journal of Climate*, 15(14), 1969-1978. Retrieved from https://
 doi.org/10.1175/1520-0442(2002)015<1969:TNTSC>2.0.CO;2
 doi: 10.1175/1520-0442(2002)015(1969:TNTSC>2.0.CO;2
- Bailey, D. A., Rhines, P. B., & Häkkinen, S. (2005, Oct 01). Formation and pathways
 of North Atlantic Deep Water in a coupled ice-ocean model of the Arctic-North
 Atlantic Oceans. *Climate Dynamics*, 25(5), 497-516. Retrieved from https://
 doi.org/10.1007/s00382-005-0050-3 doi: 10.1007/s00382-005-0050-3

518	Brambilla, E., & Talley, L. D. (2008). Subpolar Mode Water in the northeastern
519	Atlantic: 1. Averaged properties and mean circulation. Journal of Geophysical
520	Research: Oceans, 113(C4). Retrieved from https://agupubs.onlinelibrary
521	.wiley.com/doi/abs/10.1029/2006JC004062 doi: 10.1029/2006JC004062
522	Chang, P., Zhang, S., Danabasoglu, G., Yeager, S. G., Fu, H., Wang, H., Wu, L.
523	(2020). An Unprecedented Set of High-Resolution Earth System Simulations
524	for Understanding Multiscale Interactions in Climate Variability and Change.
525	<i>Journal of Advances in Modeling Earth Systems</i> , 12(12), e2020MS002298. doi:
526	https://doi.org/10.1029/2020MS002298
527	Covey, C., & Thompson, S. L. (1989). Testing the effects of ocean heat transport
528	on climate. <i>Global and Planetary Change</i> , 1(4), 331 - 341. Retrieved from
529	http://www.sciencedirect.com/science/article/pii/092181818990009X
530	doi: https://doi.org/10.1016/0921-8181(89)90009-X
531	Czaja, A., Frankignoul, C., Minobe, S., & Vannière, B. (2019). Simulating the mid-
532	latitude atmospheric circulation: what might we gain from high-resolution
533	modeling of air-sea interactions? <i>Curr. Clim. Change Rep.</i> , 5(4), 390–406.
534	Day, J. J., Hargreaves, J. C., Annan, J. D., & Abe-Ouchi, A. (2012). Sources of multi-
535	decadal variability in Arctic sea ice extent. Environmental Research Letters,
536	7(3),034011. Retrieved from http://stacks.iop.org/1748-9326/7/i=3/
537	a=034011
538	Delworth, T. L., & Zeng, F. (2016). The Impact of the North Atlantic Oscillation on
539	Climate through Its Influence on the Atlantic Meridional Overturning Circu-
540	lation. Journal of Climate, 29(3), 941-962. Retrieved from https://doi.org/
541	10.1175/JCLI-D-15-0396.1 doi: 10.1175/JCLI-D-15-0396.1
542	Delworth, T. L., Zeng, F., Vecchi, G. A., Yang, X., Zhang, L., & Zhang, R. (2016,
543	07). The North Atlantic Oscillation as a driver of rapid climate change in
544	the Northern Hemisphere. <i>Nature Geosci</i> , 9(7), 509–512. Retrieved from
545	http://dx.doi.org/10.1038/ngeo2738
546	Dong, B., & Sutton, R. T. (2003). Variability of Atlantic Ocean heat transport and its
547	effects on the atmosphere. <i>Annals of Geophysics</i> , 46(1). Retrieved from https://
548	www.annalsofgeophysics.eu/index.php/annals/article/view/3391 doi:
549	10.4401/ag-3391
550	Eden, C., & Jung, T. (2001). North Atlantic Interdecadal Variability: Oceanic Re-

551	sponse to the North Atlantic Oscillation (1865-1997). Journal of Climate, 14(5),
552	676-691. Retrieved from https://doi.org/10.1175/1520-0442(2001)
553	014<0676:NAIVOR>2.0.C0;2 doi: 10.1175/1520-0442(2001)014(0676:
554	NAIVOR>2.0.CO;2
555 Frai	nkignoul, C., Gastineau, G., & Kwon, YO. (2013). The influence of the amoc
556	variability on the atmosphere in ccsm3. <i>Journal of Climate, 26</i> (24), 9774 - 9790.
557	Retrieved from https://journals.ametsoc.org/view/journals/clim/26/
558	24/jcli-d-12-00862.1.xml doi: 10.1175/JCLI-D-12-00862.1
559 Fra	nkignoul, C., Gastineau, G., & Kwon, YO. (2015). Wintertime atmospheric re-
560	sponse to north atlantic ocean circulation variability in a climate model. Jour-
561	nal of Climate, 28(19), 7659 - 7677. Retrieved from https://journals.ametsoc
562	.org/view/journals/clim/28/19/jcli-d-15-0007.1.xml doi: 10.1175/JCLI
563	-D-15-0007.1
564 Gar	rcia-Quintana, Y., Courtois, P., Hu, X., Pennelly, C., Kieke, D., & Myers, P. G.
565	(2019). Sensitivity of Labrador Sea Water Formation to Changes in Model
566	Resolution, Atmospheric Forcing, and Freshwater Input. Journal of Geo-
567	<i>physical Research: Oceans</i> , 124(3), 2126-2152. Retrieved from https://
568	agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/2018JC014459 doi:
569	10.1029/2018JC014459
570 Gas	stineau, G., D'Andrea, F., & Frankignoul, C. (2013). Atmospheric response to the
571	north atlantic ocean variability on seasonal to decadal time scales. <i>Climate Dy-</i>
572	namics, 40(9), 2311-2330. Retrieved from https://doi.org/10.1007/s00382
573	-012-1333-0 doi: 10.1007/s00382-012-1333-0
574 Gas	stineau, G., & Frankignoul, C. (2012). Cold-season atmospheric response to
575	the natural variability of the atlantic meridional overturning circulation. <i>Cli</i> -
576	<i>mate Dynamics</i> , 39(1), 37–57. Retrieved from https://doi.org/10.1007/
577	s00382-011-1109-y doi: 10.1007/s00382-011-1109-y
578 Gri	st, J. P., Marsh, R., & Josey, S. A. (2009). On the Relationship between the North
579	Atlantic Meridional Overturning Circulation and the Surface-Forced Over-
580	turning Streamfunction. Journal of Climate, 22(19), 4989-5002. Retrieved from
581	https://doi.org/10.1175/2009JCLI2574.1 doi: 10.1175/2009JCLI2574.1
582 Hai	ada, Y., Kamahori, H., Kobayashi, C., Endo, H., Kobayashi, S., Ota, Y., Taka-
583	hashi, K. (2016). The JRA-55 Reanalysis: Representation of Atmospheric

584	Circulation and Climate Variability. Meteorological magazine. No. 2, 94(3), 269-
585	302. doi: 10.2151/jmsj.2016-015
586	Heuze, C. (2017). North Atlantic deep water formation and AMOC in CMIP5 mod-
587	els. Ocean Science, 13(4), 609-622. Retrieved from https://os.copernicus
588	.org/articles/13/609/2017/ doi: 10.5194/os-13-609-2017
589	Hurrell, J. W. (2013). The community earth system model: A framework for collabo-
590	rative research. Bull. Amer. Meteor. Soc., 94, 1339–1360.
591	Isachsen, P. E., Mauritzen, C., & Svendsen, H. (2007). Dense water formation
592	in the Nordic Seas diagnosed from sea surface buoyancy fluxes. Deep Sea
593	Research Part I: Oceanographic Research Papers, 54(1), 22-41. Retrieved from
594	http://www.sciencedirect.com/science/article/pii/S0967063706002573
595	doi: https://doi.org/10.1016/j.dsr.2006.09.008
596	Josey, S. A., Grist, J. P., & Marsh, R. (2009). Estimates of meridional overturning cir-
597	culation variability in the North Atlantic from surface density flux fields. Jour-
598	nal of Geophysical Research: Oceans, 114(C9). Retrieved from https://agupubs
599	.onlinelibrary.wiley.com/doi/abs/10.1029/2008JC005230 doi:10.1029/
600	2008JC005230
601	Kim, W. M., Yeager, S., Chang, P., & Danabasoglu, G. (2018). Low-Frequency North
602	Atlantic Climate Variability in the Community Earth System Model Large En-
603	semble. Journal of Climate, 31(2), 787-813. Retrieved from https://doi.org/
604	10.1175/JCLI-D-17-0193.1 doi: 10.1175/JCLI-D-17-0193.1
605	Kim, W. M., Yeager, S., & Danabasoglu, G. (2020). Atlantic Multidecadal Variability
606	and Associated Climate Impacts Initiated by Ocean Thermohaline Dynam-
607	ics. Journal of Climate, 33(4), 1317-1334. Retrieved from https://doi.org/
608	10.1175/JCLI-D-19-0530.1 doi: 10.1175/JCLI-D-19-0530.1
609	Kim, W. M., Yeager, S., & Danabasoglu, G. (2021). Revisiting the Causal Con-
610	nection between the Great Salinity Anomaly of the 1970s and the Shutdown
611	of Labrador Sea Deep Convection. <i>Journal of Climate</i> , 34(2), 675 - 696. Re-
612	trieved from https://journals.ametsoc.org/view/journals/clim/34/2/
613	JCLI-D-20-0327.1.xml doi: 10.1175/JCLI-D-20-0327.1
614	Kobayashi, S., Ota, Y., Harada, Y., Ebita, A., Moriya, M., Onoda, H., Takahashi, K.
615	(2015). The JRA-55 Reanalysis: General Specifications and Basic Characteris-
616	tics. Meteorological magazine. No. 2, 93(1), 5-48. doi: 10.2151/jmsj.2015-001

617	Kwon, YO., & Frankignoul, C. (2014). Mechanisms of Multidecadal Atlantic
618	Meridional Overturning Circulation Variability Diagnosed in Depth versus
619	Density Space. Journal of Climate, 27(24), 9359-9376. Retrieved from https://
620	doi.org/10.1175/JCLI-D-14-00228.1 doi: 10.1175/JCLI-D-14-00228.1
621	Langehaug, H. R., Medhaug, I., Eldevik, T., & Otterå, O. H. (2012). Arctic/atlantic
622	exchanges via the subpolar gyre. Journal of Climate, 25(7), 2421-2439. Retrieved
623	from https://doi.org/10.1175/JCLI-D-11-00085.1 doi: 10.1175/JCLI-D-11
624	-00085.1
625	Langehaug, H. R., Rhines, P. B., Eldevik, T., Mignot, J., & Lohmann, K. (2012). Wa-
626	ter mass transformation and the north atlantic current in three multicentury
627	climate model simulations. Journal of Geophysical Research: Oceans, 117(C11).
628	Retrieved from https://agupubs.onlinelibrary.wiley.com/doi/abs/
629	10.1029/2012JC008021 doi: 10.1029/2012JC008021
630	MacMartin, D. G., Zanna, L., & Tziperman, E. (2016). Suppression of Atlantic
631	Meridional Overturning Circulation Variability at Increased CO2. Journal
632	<i>of Climate</i> , 29(11), 4155-4164. Retrieved from https://doi.org/10.1175/
633	JCLI-D-15-0533.1 doi: 10.1175/JCLI-D-15-0533.1
634	Marsh, R. (2000). Recent Variability of the North Atlantic Thermohaline Circulation
635	Inferred from Surface Heat and Freshwater Fluxes. Journal of Climate, 13(18),
636	3239-3260. Retrieved from https://doi.org/10.1175/1520-0442(2000)
637	013<3239:RV0TNA>2.0.C0;2 doi: 10.1175/1520-0442(2000)013(3239:
638	RVOTNA>2.0.CO;2
639	McCartney, M. S., & Talley, L. D. (1982). The Subpolar Mode Water of the North
640	Atlantic Ocean. Journal of Physical Oceanography, 12(11), 1169-1188. Re-
641	trieved from https://doi.org/10.1175/1520-0485(1982)012<1169:
642	TSMWOT>2.0.C0;2 doi: 10.1175/1520-0485(1982)012(1169:TSMWOT)2.0.CO;2
643	Mecking, J. V., Keenlyside, N. S., & Greatbatch, R. J. (2015, Sep 01). Multiple
644	timescales of stochastically forced North Atlantic Ocean variability: A model
645	study. Ocean Dynamics, 65(9), 1367–1381. Retrieved from https://doi.org/
646	10.1007/s10236-015-0868-0 doi: 10.1007/s10236-015-0868-0
647	Menary, M. B., Hodson, D. L. R., Robson, J. I., Sutton, R. T., Wood, R. A., & Hunt,
648	J. A. (2015). Exploring the impact of CMIP5 model biases on the simula-
649	tion of North Atlantic decadal variability. Geophysical Research Letters, 42(14),

650	5926-5934. Retrieved from http://dx.doi.org/10.1002/2015GL064360
651	(2015GL064360) doi: 10.1002/2015GL064360
652	Newsom, E. R., Bitz, C. M., Bryan, F. O., Abernathey, R., & Gent, P. R. (2016). South-
653	ern Ocean Deep Circulation and Heat Uptake in a High-Resolution Climate
654	Model. Journal of Climate, 29(7), 2597-2619. Retrieved from https://doi.org/
655	10.1175/JCLI-D-15-0513.1 doi: 10.1175/JCLI-D-15-0513.1
656	Oldenburg, D., Armour, K. C., Thompson, L., & Bitz, C. M. (2018). Distinct Mecha-
657	nisms of Ocean Heat Transport Into the Arctic Under Internal Variability and
658	Climate Change. Geophysical Research Letters, 45(15), 7692-7700. Retrieved
659	<pre>from https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/</pre>
660	2018GL078719 doi: 10.1029/2018GL078719
661	Oldenburg, D., Wills, R. C. J., Armour, K. C., Thompson, L., & Jackson, L. C. (2021).
662	Mechanisms of Low-Frequency Variability in North Atlantic Ocean Heat
663	Transport and AMOC. <i>Journal of Climate</i> , 34(12), 4733-4755. Retrieved
664	from https://journals.ametsoc.org/view/journals/clim/34/12/
665	JCLI-D-20-0614.1.xml doi: 10.1175/JCLI-D-20-0614.1
666	Pérez-Brunius, P., Rossby, T., & Watts, D. R. (2004). Transformation of the Warm
667	Waters of the North Atlantic from a Geostrophic Streamfunction Perspec-
668	tive. Journal of Physical Oceanography, 34(10), 2238-2256. Retrieved from
669	https://doi.org/10.1175/1520-0485(2004)034<2238:T0TWWO>2.0.C0;2
670	doi: 10.1175/1520-0485(2004)034(2238:TOTWWO)2.0.CO;2
671	Pickart, R. S., & Spall, M. A. (2007). Impact of Labrador Sea Convection on the
672	North Atlantic Meridional Overturning Circulation. Journal of Physical
673	Oceanography, 37(9), 2207-2227. Retrieved from https://doi.org/10.1175/
674	JP03178.1 doi: 10.1175/JPO3178.1
675	Rahmstorf, S. (2002). Ocean circulation and climate during the past 120,000 years.
676	<i>Nature</i> , 419(6903), 207–214. Retrieved from https://doi.org/10.1038/
677	nature01090 doi: 10.1038/nature01090
678	Roberts, C. D., Waters, J., Peterson, K. A., Palmer, M. D., McCarthy, G. D., Frajka-
679	Williams, E., Zuo, H. (2013). Atmosphere drives recent interannual vari-
680	ability of the Atlantic meridional overturning circulation at 26.5 °N. <i>Geophys</i> -
681	ical Research Letters, 40(19), 5164-5170. Retrieved from http://dx.doi.org/
682	10.1002/grl.50930 doi: 10.1002/grl.50930

683	Robson, J., Ortega, P., & Sutton, R. (2016). A reversal of climatic trends in the north
684	atlantic since 2005. <i>Nature Geoscience</i> , 9(7), 513–517. Retrieved from https://
685	doi.org/10.1038/ngeo2727 doi: 10.1038/ngeo2727
686	Sein, D. V., Koldunov, N. V., Danilov, S., Sidorenko, D., Wekerle, C., Cabos, W.,
687	Jung, T. (2018). The Relative Influence of Atmospheric and Oceanic Model
688	Resolution on the Circulation of the North Atlantic Ocean in a Coupled Cli-
689	mate Model. Journal of Advances in Modeling Earth Systems, 10(8), 2026-2041.
690	Retrieved from https://agupubs.onlinelibrary.wiley.com/doi/abs/
691	10.1029/2018MS001327 doi: 10.1029/2018MS001327
692	Speer, K., & Tziperman, E. (1992). Rates of Water Mass Formation in the North
693	Atlantic Ocean. Journal of Physical Oceanography, 22(1), 93-104. Retrieved from
694	https://doi.org/10.1175/1520-0485(1992)022<0093:ROWMFI>2.0.CO;2
695	doi: 10.1175/1520-0485(1992)022(0093:ROWMFI)2.0.CO;2
696	Straneo, F. (2006). On the Connection between Dense Water Formation, Overturn-
697	ing, and Poleward Heat Transport in a Convective Basin. Journal of Physical
698	Oceanography, 36(9), 1822-1840. Retrieved from https://doi.org/10.1175/
699	JP02932.1 doi: 10.1175/JPO2932.1
700	Treguier, A. M., Theetten, S., Chassignet, E. P., Penduff, T., Smith, R., Talley, L.,
701	Böning, C. (2005). The North Atlantic Subpolar Gyre in Four High-Resolution
702	Models. Journal of Physical Oceanography, 35(5), 757-774. Retrieved from
703	https://doi.org/10.1175/JP02720.1 doi: 10.1175/JPO2720.1
704	Tziperman, E. (1986). On the Role of Interior Mixing and Air-Sea Fluxes
705	in Determining the Stratification and Circulation of the Oceans. Jour-
706	nal of Physical Oceanography, 16(4), 680-693. Retrieved from https://
707	doi.org/10.1175/1520-0485(1986)016<0680:0TR0IM>2.0.CO;2 doi:
708	10.1175/1520-0485(1986)016(0680:OTROIM)2.0.CO;2
709	Walin, G. (1982). On the relation between sea-surface heat flow and thermal
710	circulation in the ocean. <i>Tellus</i> , <i>34</i> (2), 187-195. Retrieved from https://
711	onlinelibrary.wiley.com/doi/abs/10.1111/j.2153-3490.1982.tb01806.x
712	doi: 10.1111/j.2153-3490.1982.tb01806.x
713	Wen, N., Frankignoul, C., & Gastineau, G. (2016, Oct 01). Active AMOC-NAO
714	coupling in the IPSL-CM5A-MR climate model. Climate Dynamics, 47(7), 2105-
715	2119. Retrieved from https://doi.org/10.1007/s00382-015-2953-y doi: 10

716	.1007/s00382-015-2953-y
717	Wills, R. C., Schneider, T., Wallace, J. M., Battisti, D. S., & Hartmann, D. L. (2018).
718	Disentangling Global Warming, Multidecadal Variability, and El Niño in Pa-
719	cific Temperatures. Geophysical Research Letters, 45(5), 2487-2496. Retrieved
720	<pre>from https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1002/</pre>
721	2017GL076327 doi: 10.1002/2017GL076327
722	Wills, R. C. J., Armour, K. C., Battisti, D. S., & Hartmann, D. L. (2019). Ocean-
723	Atmosphere Dynamical Coupling Fundamental to the Atlantic Multidecadal
724	Oscillation. <i>Journal of Climate</i> , 32(1), 251-272. Retrieved from https://
725	doi.org/10.1175/JCLI-D-18-0269.1 doi: 10.1175/JCLI-D-18-0269.1
726	Winton, W. A. T. D. S. G. W. H A. R., M. (2014). Has coarse ocean resolution bi-
727	ased simulations of transient climate sensitivity? Geophysical Research Letters,
728	41, 8522-8529.
729	Zhang, R. (2015). Mechanisms for low-frequency variability of summer Arctic sea
730	ice extent. Proceedings of the National Academy of Sciences, 112(15), 4570-4575.
731	Retrieved from http://www.pnas.org/content/112/15/4570.abstract doi:
732	10.1073/pnas.1422296112

Supporting Information for "AMOC and water-mass transformation in high- and low-resolution models: Climatology and variability"

Dylan Oldenburg¹, Robert Jnglin Wills², Kyle C. Armour^{1,2}, LuAnne

 $Thompson^1$

 $^1{\rm School}$ of Oceanography, University of Washington, Seattle, Washington

²Department of Atmospheric Sciences, University of Washington, Seattle, Washington

Contents of this file

1. Figure S1: Colors: Total climatological winter surface density flux calculated using the methodology from Oldenburg et al. (2021) over densities less than the density where AMOC reaches 75% of its maximum. Contours: Time-mean winter sea-surface potential density referenced to 2000 m for **a** JRA55-HR, **b** JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

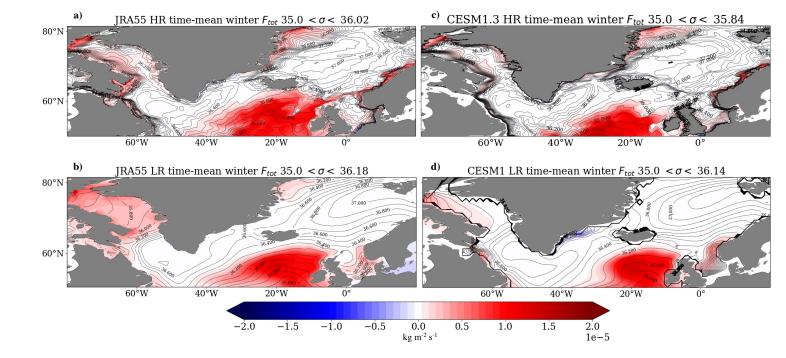
2. Figure S2: **a**) JRA55-HR total surface freshwater flux climatology. **b-d**) Total surface freshwater flux climatologies (contours) and anomalies relative to JRA55-HR (colors) for **b**) JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

X - 2

3. Figure S3: Lead-lag regressions of annual-mean AMOC in density coordinates onto the first LFC of AMOC σ for (a-f) CESM1-HR and (g-l) CESM1-LR. Lead times indicate anomalies that lead the LFC, i.e., prior to the time of maximum AMOC. Because the LFCs are unitless, the regressions simply have units of Sv.

:

4. Figure S4: Lead-lag regressions of mixed-layer depth averaged over January, February and March onto the first LFC of AMOC σ for (a-f) CESM1-HR and (g-l) CESM1-LR. Lead times indicate anomalies that lead the LFC, i.e., prior to the time of maximum AMOC. Because the LFCs are unitless, the regressions simply have units of m.



:

Figure S1. Colors: Total climatological winter surface density flux calculated using the methodology from Oldenburg et al. (2021) over densities less than the density where AMOC reaches 75% of its maximum. Contours: Time-mean winter sea-surface potential density referenced to 2000 m for **a** JRA55-HR, **b** JRA55-LR, **c**) CESM1-HR and **d**) CESM1-LR.

X - 4

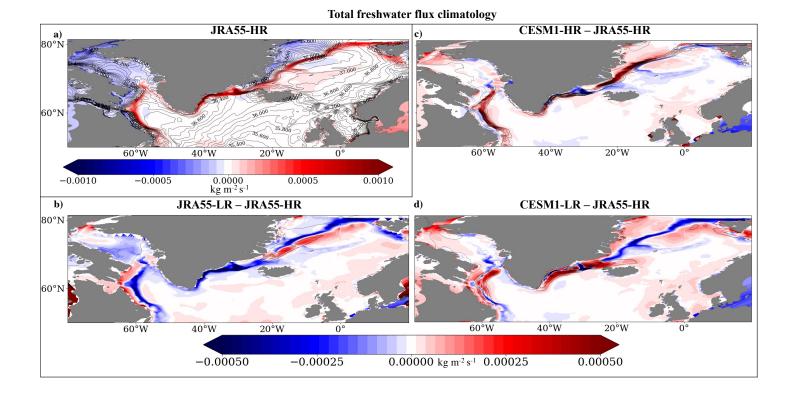


Figure S2. a) JRA55-HR total surface freshwater flux climatology. b-d) Total surface freshwater flux climatologies (contours) and anomalies relative to JRA55-HR (colors) for b) JRA55-LR, c) CESM1-HR and d) CESM1-LR.

February 15, 2022, 3:12pm

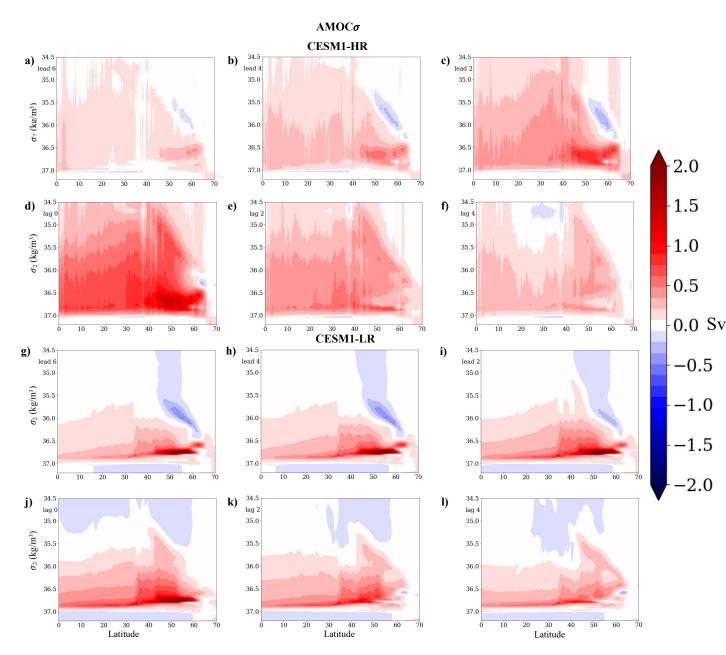
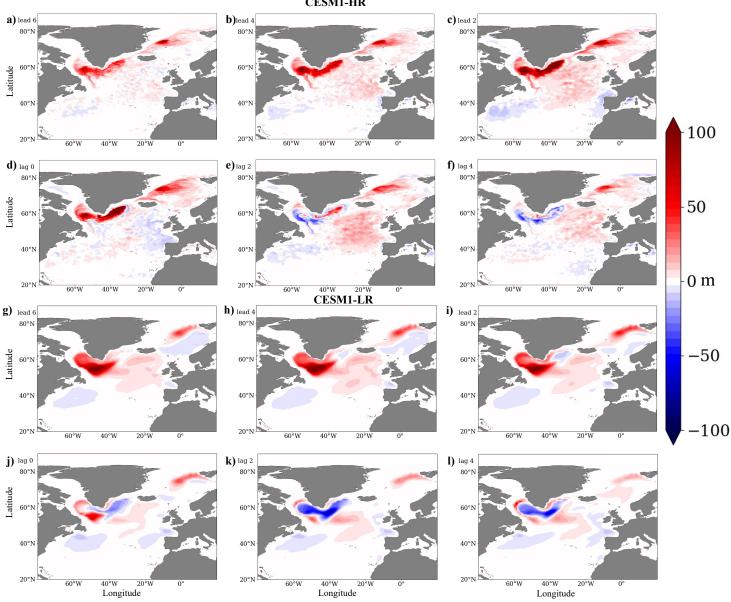


Figure S3. Lead-lag regressions of annual-mean AMOC in density coordinates onto the first LFC of AMOC σ for (a-f) CESM1-HR and (g-l) CESM1-LR. Lead times indicate anomalies that lead the LFC, i.e., prior to the time of maximum AMOC. Because the LFCs are unitless, the regressions simply have units of Sv.

February 15, 2022, 3:12pm



Winter mixed-layer depth CESM1-HR

:

Figure S4. Lead-lag regressions of mixed-layer depth averaged over January, February and March onto the first LFC of AMOC σ for (a-f) CESM1-HR and (g-l) CESM1-LR. Lead times indicate anomalies that lead the LFC, i.e., prior to the time of maximum AMOC. Because the LFCs are unitless, the regressions simply have units of m.

February 15, 2022, 3:12pm